

SIPCO2: A simple, inexpensive surface water pCO₂ sensor

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Abstract

Efforts to estimate air-water carbon dioxide (CO₂) exchange on regional or global scales are constrained by a lack of direct, continuous surface water CO₂ observations. Sensor technology for the in situ measurement of the partial pressure of carbon dioxide (pCO₂) has progressed, but still poses limitations including expense and bio-fouling concerns. We describe a simple, inexpensive, in situ pCO₂ method (SIPCO2) in which a non-dispersive infrared (NDIR) detector is paired with an air pump in an enclosed housing to produce air-sea equilibration. We first evaluated this approach in a laboratory setting, then in an estuarine-coastal ocean laboratory for several months to continuously monitor aquatic pCO₂. An accepted, accurate NDIR-based CO₂ measurement technique was employed alongside SIPCO2 to provide an assessment of sensor performance. SIPCO2 allows for low-cost, relatively accurate measurements of pCO₂ (mean difference of -5 ± 5 μatm from validation system after laboratory calibration) without reagents or membranes, and can be assembled and operated with a minimal amount of technical skill. While not suitable for some exacting applications, this SIPCO2 approach could rapidly and effectively increase the number of quality CO₂ observations in a range of aquatic environments. We also provide detailed instructions for the assembly of SIPCO2 from commercially available components.

Recent examinations of the fluxes of carbon dioxide (CO₂) between the terrestrial, atmospheric, inland aquatic, estuary, and coastal ocean provinces have illustrated both their dynamic nature and importance to regional and global carbon budgets (Frankignoulle 1988; Raymond et al. 1997; Cole et al. 2007; Raymond et al. 2013; Signorini et al. 2013; Laruelle et al. 2014, 2015; Borges et al. 2015). Many of these studies, particularly those focused on inland waters, have not used direct observations of the partial pressure of carbon dioxide (pCO₂) in water, but instead have derived pCO₂ from other measures of the carbonate system for some or all of the analysis (Raymond et al. 1997; Cole et al. 2007; Atilla et al. 2011; Butman and Raymond 2011; McDonald et al. 2013; Laruelle et al. 2015). This derivation is commonly accomplished through the pairing of discrete total alkalinity and pH sample data (Park 1969), a technique which has been shown to result in a large potential overestimate of pCO₂ due to diverse factors such as interferences (particularly in freshwater systems) from non-carbonate contributions to alkalinity (Hunt et al. 2011; Wang et al. 2013; Abril et al. 2015) and electrode-based pH measurement errors in seawater (Hoppe et al. 2012; Bockmon and Dickson 2015).

While a more complete understanding of non-carbonate contributions to alkalinity could potentially lead to accurate indirect estimates of pCO₂ from discrete measurements—as could the pairing of sample measurements of dissolved inorganic carbon and pH—an alternative way to improve and constrain CO₂ budgets is to vastly increase the number of direct pCO₂ observations, especially in freshwaters where non-carbonate interferences are the most prevalent. The need for direct pCO₂ measurements, together with technological improvements, has resulted in a variety of innovative new pCO₂ sensor technologies. These technologies employ several equilibration techniques, including headspace equilibration through a gas-permeable membrane and non-membrane equilibration, as well as contrasting CO₂ detection techniques such as non-dispersive infrared (NDIR) detection and chemical reagent detection (Table 1). An evaluation of four commercial pCO₂ sensors in coastal waters by the Alliance for Coastal Technology in 2010 found a range of sensor performance compared to a documented underway-type pCO₂ system (using a membrane-based equilibrators and NDIR CO₂ detection, Hales and Takahashi 2004) and discrete pCO₂ measurements (pCO₂ derived from total alkalinity and pH). A separate evaluation of a membrane-equilibration technique compared sensor data to discrete headspace CO₂ measurements with a large range of reported sensor discrepancy (Johnson et al. 2010).

Although these commercial in-water measurement sensors are generally pre-configured for immediate use, they are also

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Additional Supporting Information may be found in the online version of this article.

Table 1. Aquatic CO₂ sensors and their performance characteristics.

Manufacturer/ sensor	Measurement method	Range of difference from standard system (mean difference)	Standard deviation range	Citation
Contros Hydro-C	Pumped membrane equilibration, NDIR detection	-16 to +96 μ atm	+/- 17 to 25 μ atm	Alliance for Coastal Technologies (ACT) (2010a)
Pro-Oceanus CO ₂ -Pro	Pumped membrane equilibration, NDIR detection	+9 μ atm	+/- 14 μ atm	Alliance for Coastal Technologies (ACT) (2010b)
Vaisala CARBOCAP (modified)	Passive membrane equilibration, NDIR detection	+5 to +881 ppm*	Not reported	Johnson et al. (2010)
Sunburst SAMI-CO ₂	Passive membrane equilibration, reagent-based colorimetry	+18 to +40 μ atm	+/- 9 to 40 μ atm	Alliance for Coastal Technologies (ACT) (2010c)
PMEL/Batelle MAPCO ₂	Bubble equilibration, NDIR detection	-12 to +3 μ atm	+/- 8 to 30 μ atm	Alliance for Coastal Technologies (ACT) (2010d)
This study- before characterization	Pumped air equilibration, NDIR detection	+7 to +65 μ atm†	+/- 6 μ atm	This study
This study- after characterization	Pumped air equilibration, NDIR detection	-26 to +30 μ atm‡	+/- 6 μ atm	This study

* Converted from mg/L, assuming 0°C and 1.01325 bar pressure.

† SIPCO₂ CML data without laboratory calibration applied.

‡ SIPCO₂ CML data with laboratory calibration applied.

vulnerable to biofouling (especially in the case of membrane-type systems and the SAMI-CO₂ water intake tube), are sometimes complex to install due to multiple components requiring specialized mounting hardware (MAPCO₂), and costly. Recently, Bastviken et al. (2015) presented a innovative design for a passively equilibrated pCO₂ sensor, using the same low-cost CO₂ detector employed in this study, with good detector linearity and promising results against discrete bottle-headspace pCO₂ measurements. These results indicate that these inexpensive detectors hold great potential for the construction of an in situ pCO₂ sensor. However, the bottle-headspace method has several shortcomings (Bastviken et al. 2015), and that study did not provide assessment of performance against a validated continuous headspace-equilibration system. Here, we present an alternative design for a simple, inexpensive pCO₂ sensor (SIPCO₂) that can be assembled in a few hours from off-the-shelf or easily obtained components totaling less than \$500 USD. We then present results from two evaluation exercises, the first a multi-day lab test in a tank of manipulated freshwater, the second a multi-month test at a coastal laboratory. While we emphasize and discuss how evaluation conditions differed between the SIPCO₂ evaluations and those of commercially available sensors, and thus results may not be directly comparable, SIPCO₂ performance was similar to that of the sensors listed in Table 1 once calibrated against an accepted pCO₂ measurement system.

Materials and procedures

Equilibration background

Among all styles of in-water CO₂ sensors, one common element is the necessity to separate aqueous CO₂ from the

liquid medium into a measureable state. All the sensors in Table 1 measure CO₂ in the gaseous state via NDIR detection with the exception of the Sunburst SAMI-CO₂, which measures pH change in the liquid phase as CO₂ exchanges across a membrane (DeGrandpre et al. 1995). In marine pCO₂ observations the technique of equilibrating an air headspace with a water mass of interest has been used for decades (Takahashi 1961; Weiss 1981; Wanninkhof and Thoning 1993). According to this technique, after equilibration the gas composition of the headspace reflects that of the water mass. Often the equilibrator is a showerhead-type in which the air headspace is ultimately pumped to a NDIR detector at a set rate. The MAPCO₂ system achieves equilibration through a combination of wave and bubble action, and components in contact with seawater are constructed of copper alloy to prevent biofouling (Friederich et al. 1995, ACT 2010). All types of equilibrators can be vulnerable to fouling, even those constructed using an anti-fouling agent such a copper, and must be cleaned often to maintain accurate pCO₂ measurements. More recently, use of this technique has migrated upstream to estuary and freshwater environments (Frankignoulle and Borges 2001; Abril et al. 2006). The underlying principle of this type of equilibration requires that a relatively large volume of water is exchanged with a relatively small air headspace to equilibrate the CO₂ levels between the two media. This concept requires the transfer via pumps of small volumes of air and larger volumes of water. Our design simplifies this concept by recirculating headspace air below the water surface via a small air pump; the resulting bubbles equilibrate CO₂ with the surrounding water, break, and the released air rises through

tubing or a pipe back to the headspace containing a small NDIR detector (Fig. 1). The same headspace air is recirculated, providing equilibration at a rate suitable for applications where: (1) the sample water is not rapidly changing its carbonate characteristics, and (2) bubbles do not escape the return pipe or the enclosure is not opened to ambient air.

SIPCO₂ sensor design

The SIPCO₂ sensor design (Fig. 1) has two electro-mechanical components—the air pump and NDIR detector—housed in an off-the-shelf plastic electrical junction box (component parts list included in Supporting Information Material). In order to maintain equilibration, the air pump is run continuously, although it may be possible to run the pump intermittently to conserve power, especially when powered by a battery. The flow rate of the selected pump is listed by the manufacturer (KNF Neuberger) as 0.45 L min⁻¹, with an operating life of 10,000 h for the brushless model, although in other applications we have observed pump durability of 4–5 yr.

The NDIR detector employed in this work was a SenseAir K-30 (ASCII model, SenseAir AB, Delsbo Sweden). The K-30 detector has a manufacturer-specified measurement range of 0–10,000 ppm CO₂, an accuracy of ± 30 ppm $\pm 3\%$ of the measurement value, and a precision of ± 20 ppm $\pm 1\%$ of the measurement value. Our own results found that with calibration the K-30 can achieve both better accuracy and better precision, as discussed below. The ASCII output version of the K-30 sensor used in this work provides an automatic ASCII-format CO₂ measurement every 2 s. A vendor-supplied control program (“Gaslab”) was used to calibrate the K-30, although this program is limited to calibration concentrations of 0 and 400 ppm only. For both the laboratory evaluation and coastal laboratory evaluation discussed below, the SIPCO₂ sensor was powered by a standard 5-volt Universal Serial Bus connection to a laptop computer. This computer also logged the ASCII SIPCO₂ data output via a terminal program. The K-30 comes equipped with an onboard temperature sensor. While we performed a later evaluation of the temperature dependency of the K-30 (described below), the onboard temperature component failed during the two central evaluations of this work.

Validation system

The validation system, typically used for shipboard continuous-flow pCO₂ measurements (see Hunt et al. 2013) consisted of an equilibrator with spray nozzle, with headspace air recirculated by a small air pump between the equilibrator and a Li-cor LI-840 NDIR detector. The manufacturer-specified operating range of the LI-840 NDIR is 0–3000 ppm CO₂. The validation system was calibrated with ultrapure nitrogen and an 835 ppm CO₂/nitrogen mixture, and a computer-controlled system of valves flowed the same calibration gases through the validation LI-840 NDIR once every

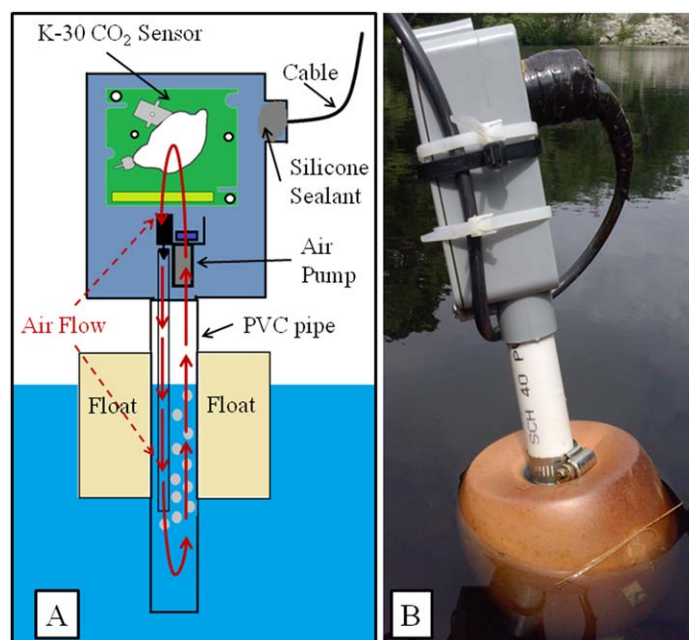


Fig. 1. Panel **A**: SIPCO₂ sensor design. A components list is provided in the Supporting Information Material. Panel **B**: the prototype SIPCO₂ sensor deployed in the Lamprey River, New Hampshire U.S.A.

3 h to check system calibration stability. The estimated uncertainty of the validation system is ± 3 μ atm (Vandemark et al. 2011).

pCO₂ calculation

The SIPCO₂ K-30 detector reported measurements as the molar fraction of CO₂ (x_{CO_2}). Since the sensor housing was open to the water surface, the readings reported by the SIPCO₂ sensor were “wet” ($x_{CO_{2wet}}$). These $x_{CO_{2wet}}$ data were corrected to sea surface temperature, assuming the temperature inside the SIPCO₂ was also equilibrated to the observed sea surface temperature, and converted from $x_{CO_{2wet}}$ to the partial pressure of carbon dioxide (pCO₂, wet air), according to standard methods (Dickson et al. 2007), assuming the SIPCO₂ headspace was at ambient atmospheric pressure, measured by a nearby meteorological array. In contrast, the sample stream of the validation system described above was drawn through tubing containing a Nafion selectively permeable membrane (Perma Pure, Toms River, New Jersey), which dried the sample gas stream of water vapor, resulting in validation system readings of $x_{CO_{2dry}}$. These $x_{CO_{2dry}}$ validation data were corrected to sea surface temperature and converted from $x_{CO_{2dry}}$ to pCO₂ as above (Dickson et al. 2007). Thus, all pCO₂ data presented in this work are in the form of partial pressures of CO₂ in wet air.

Laboratory evaluation

The K-30 detector was initially removed from the SIPCO₂ housing and calibrated in the lab by placing the sensor in a

sealed plastic bag with a small opening for gas tubing and detector cable. When gas flow was introduced the positive pressure flushed the bag after several minutes, allowing for stable CO₂ readings from the detector. The detector was zeroed with ultrapure nitrogen, then subsequently spanned with a 400 ppm CO₂/nitrogen mixture. The limitations of this detector calibration procedure are discussed later in this work. After calibration, the K-30 detector was placed into the sensor housing and run in the lab continuously for several days alongside a more traditional aquatic CO₂ system. Both pCO₂ systems measured a tank of ~ 200 L tap water, with the SIPCO2 sensor placed directly into the tank, while water was recirculated between the tank and validation system's equilibrator via a submersible pump. The temperature of the SIPCO2 headspace was assumed to be equal to that of the water in the tank. The simultaneous use of the SIPCO2 and validation systems allowed us to compare the behavior of the SIPCO2 sensor over a wider range of CO₂ in the lab, as carbonated water was added to the tank twice to raise CO₂ levels (Fig. 2). These carbonated water additions also allowed the estimation of the SIPCO2 sensor response time.

Coastal marine laboratory evaluation (CML)

A longer-term sensor evaluation was conducted from 22 November 2014 to 12 March 2015 at the University of New Hampshire Coastal Marine Laboratory (CML), a facility at the mouth of the Piscataqua River, which drains to the coastal Gulf of Maine (43.0°N 70.7°W). The facility has a continuous seawater supply, pumped from near the bottom of the adjacent channel at a tide-varying depth between 4.5 m and 7 m. The same sensor used in the laboratory evaluation, which was not recalibrated following the laboratory evaluation, was placed into a ~ 200 L tank similar to that used in the laboratory evaluation, which was continuously flushed with seawater at an approximate rate of 10 L/min. The CML tank also contained a thermosalinograph (Aanderaa 4319B), which collected data for 10 min each hour. Salinities reported in this paper are on the PSS-78 scale (Lewis 1980). A separate seawater line continually supplied the headspace gas equilibrator for the CML CO₂ seawater measurement system. This system was identical to that used in the laboratory evaluation, and measurements of ultrapure nitrogen and an 835 ppm CO₂/nitrogen mixture were made every 4 h to determine the stability of the validation system. While the SIPCO2 and validation CO₂ systems were run continuously, only the pCO₂ data from these systems collected over the 10-min period when the thermosalinograph was in operation were used to produce 10-min average values once per hour, which comprise the dataset presented here.

Assessment

Laboratory evaluation

During the laboratory evaluation, the SIPCO2 sensor closely tracked the validation system during unperturbed

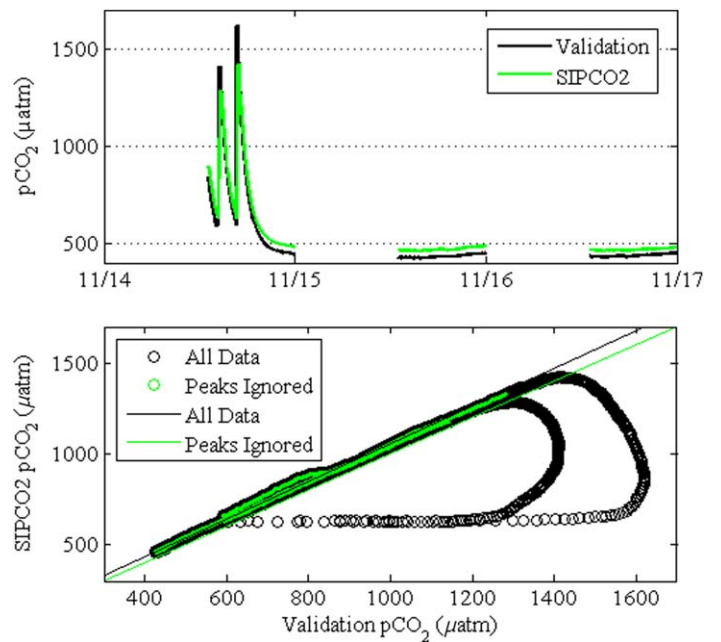


Fig. 2. Laboratory SIPCO2 sensor evaluation. Top panel: timeseries plot of SIPCO2 and validation system pCO₂ measurements. Difficulties with the K-30 CO₂ sensor logging software, which reset at the end of each day, resulted in the data gaps shown. Bottom panel: scatterplot of SIPCO2 pCO₂ readings against validation system pCO₂ readings, including two CO₂ additions (green points) and excluding CO₂ additions (black points).

periods ($r^2 = 0.99$). Shortly after two rapid additions of CO₂ the SIPCO2 sensor response lagged that of the validation system (Fig. 2). Results of mean offsets (Off_{mean} , calculated as differences between sensor readings, Eq. 1) show that readings from the SIPCO2 sensor were consistently higher than those from the validation system (excepting the periods of rapid CO₂ increase during the two perturbation periods). This consistent offset was present despite the calibration of the SIPCO2 NDIR detector with the same gas standards used to calibrate the validation system (Table 2).

$$Off_{mean} = \frac{1}{n} * \sum_{i=1}^n (pCO_{2 \text{ SIPCO2}} - pCO_{2 \text{ Validation}}) \quad (1)$$

There is a substantial discrepancy between the mean offsets if data from the perturbation periods are included (Table 2). Standard deviations of the mean offsets (calculated as in Eq. 2) show that the perturbation periods represented a large amount of variance, while ignoring these periods yielded a standard deviation (precision) of ± 14 μatm. A ± 14 μatm precision, which translates to a ± 3.5% relative uncertainty at a pCO₂ of 400 μatm, represents an unacceptable level of uncertainty for exacting applications; for instance, this uncertainty exceeds the desired levels of both the “weather” and “climate” objectives (2.5% and 0.5% desired

Table 2. Offset and linear regression results of the laboratory evaluation of the SIPCO2 sensor.

	With peaks	Without peaks
Off _{mean} (μatm)	30.2 (±76.7)	42.7 (±14.2)
Slope	0.88	1.03
Intercept	105	22
R ²	0.92	0.99

uncertainty, respectively) of the Global Ocean Acidification Observing Network (GOA-ON, Newton et al. 2015).

$$\text{Off}_{\text{std}} = \sqrt{\frac{1}{n} \sum_{i=1}^n (\text{pCO}_2_{\text{SIPCO2}} - \text{pCO}_2_{\text{validation}})^2} \quad (2)$$

Linear regressions of the SIPCO2 sensor readings against those from the validation system, both for the entire comparison period and for the two periods without CO₂ additions are listed in Table 2. The response time of the SIPCO2 system lagged that of the validation system, resulting in the separation of the two peaks in Fig. 2 (top panel) and deviation from linearity apparent in Fig. 2 (bottom panel). The time lag between the highest validation system pCO₂ and SIPCO2 pCO₂ was approximately 9 min for the first peak, and 11 min for the second peak, indicating a slower response time for the SIPCO2 sensor. The validation system responded quickly to the CO₂ additions (less than one minute from initial addition to the start of the validation system response), while 3 min and 4 min elapsed between the start of the validation response to peak validation pCO₂, resulting in total validation system response times of 4 and 5 min for the first and second peaks, respectively. Addition of these validation system response times with the 9 min and 11 min lags between the validation and SIPCO2 peaks, together with a 1 min lag between the CO₂ addition and start of validation system response, results in a conservative SIPCO2 response time estimate of 15 min. The highest SIPCO2 readings for the two peaks (1284 μatm and 1421 μatm, respectively) were lower than those from the validation system (1411 μatm and 1618 μatm, respectively), suggesting that the added CO₂ from the carbonated water escaped the tank quickly enough that the SIPCO2 sensor could not fully equilibrate due to the slower SIPCO2 sensor response time. This is probably due in part to the rapid recirculation of water through the validation system’s equilibrator; thus, we might expect a slower CO₂ decrease in naturally less-turbulent waters.

We performed a linear regression of the SIPCO2 sensor readings against those from the validation system to examine the linear response of the SIPCO2 sensor (Table 2; Fig. 2). When the peaks from the CO₂ additions were excluded,

Table 3. Summary of hydrographic conditions and pCO₂ observations from the Coastal Marine Lab evaluation. Min is the minimum measurement during the evaluation, Max is the maximum measurement observed, SD is the standard deviation of the measurement during the evaluation (n = 2,539 hourly means).

	Mean	Median	Min	Max	SD
Salinity	29.95	30.45	22.31	32.28	1.87
Water temperature	4.31	4.85	-0.75	10.39	2.93
Dissolved oxygen (μmol l ⁻¹)	345	338	269	433	33
Validation pCO ₂	402	399	338	472	27
K-30 pCO ₂	431	430	360	501	30

the SIPCO2 sensor response was quite linear to that of the validation system (r² = 0.99), with a near 1 : 1 slope.

Coastal marine laboratory evaluation

Hydrographic conditions at the CML varied over the approximately three and a half months the SIPCO2 sensor was deployed (Table 3). Water temperatures gradually decreased from 10°C to near 0°C as the season progressed from late fall to winter, then began to rise at the end of the evaluation in spring. Salinity reached a minimum in late December due to higher than normal rainfall. Underlying these seasonal changes were clear semi-diurnal tidal signals, with high tide bringing warmer water temperatures and higher salinities. The magnitudes of salinity and temperature changes varied with lunar phase, as well as with precipitation and associated changes in freshwater input, but water temperature generally varied 1–2°C between low and high tides, while salinity varied 1.5–3.

As in the laboratory evaluation, the SIPCO2 sensor sensitively tracked the readings of the validation pCO₂ system at the CML, but with somewhat higher readings (Fig. 3; Tables 2, 4). The offset between the SIPCO2 sensor and validation system (Fig. 3, bottom panel) varied from 7 μatm to 65 μatm over the CML test, with a mean offset of 29 μatm and standard deviation of 6 μatm (Table 4). This mean offset from the CML evaluation is somewhat lower than the mean offset found in the laboratory evaluation (42.7 μatm, without peaks), but within one standard deviation of the laboratory evaluation mean (Table 2). Paired t-tests show that the mean CML offset is significantly different than the mean laboratory evaluation offset at the 95% significance level. The results of a linear regression of the SIPCO2 readings against those from the validation system at the CML yielded slightly different slopes than observed in the laboratory evaluation (1.08 μatm and 1.03 μatm, respectively), as well as different intercepts (-1.9 μatm and 22 μatm, respectively).

Coastal marine laboratory evaluation data correction

We used the results of the laboratory evaluation results in two ways to attempt to correct the SIPCO2 sensor readings

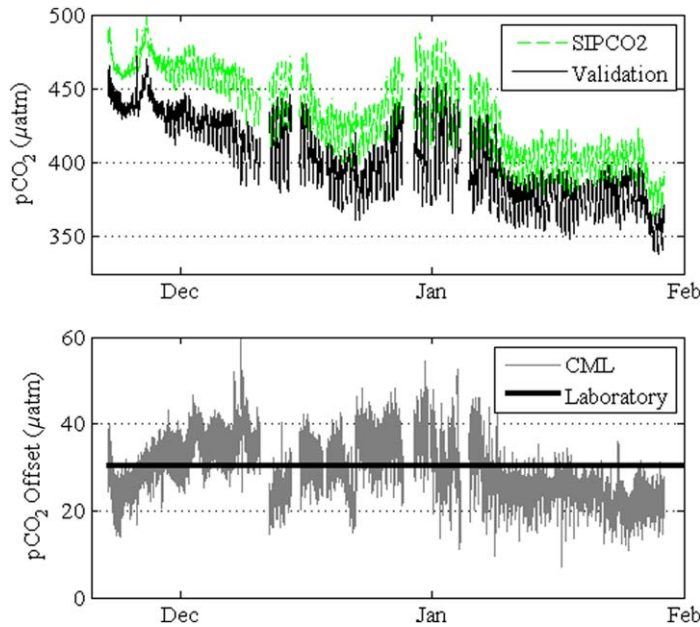


Fig. 3. Comparison data collected at the UNH Coastal Marine Lab. Top panel: time series of pCO₂ readings from SIPCO2 and validation systems. Data gaps in this panel are due to the removal of data when the validation system equilibrator was clogged. Bottom panel: time series of sensor offsets, with the mean offset from the laboratory evaluation shown as a solid line. Offsets were calculated as the difference between SIPCO2 and validation system measurements.

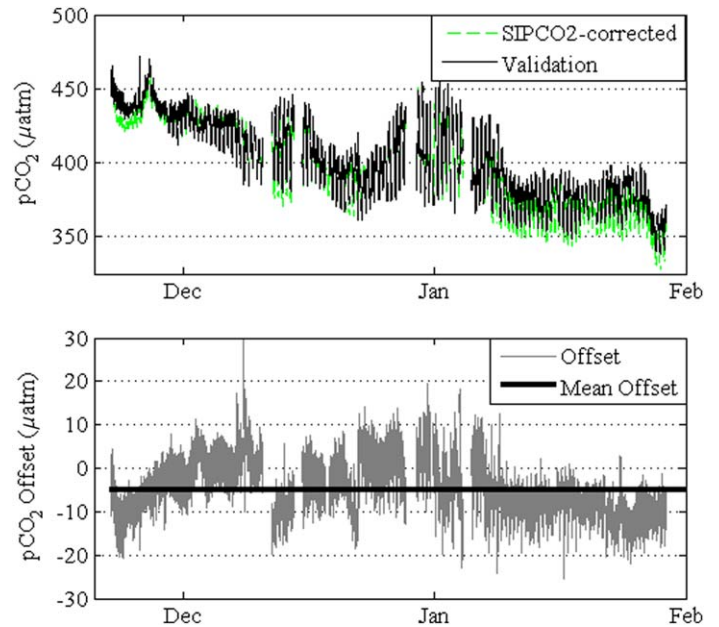


Fig. 4. Corrected comparison data collected at the UNH Coastal Marine Lab. Top panel: time series of corrected SIPCO2 sensor and validation system readings. The SIPCO2 correction was derived from the laboratory evaluation results (Eq. 3). Bottom panel: time series of sensor offsets, with the mean offset from the coastal laboratory evaluation ($-5 \mu\text{atm}$) shown as a solid line. Linear regression of the corrected CML SIPCO2 sensor data against validation system readings results in an equation of $p\text{CO}_2 \text{ SIPCO2} = 1.08 * p\text{CO}_2 \text{ validation} - 30.8$, with a r^2 of 0.969 and $p \ll 0.001$.

Table 4. Offsets and regressions of uncorrected and corrected data from the CML SIPCO2 sensor test. Offsets were calculated as $p\text{CO}_2 \text{ SIPCO2} - p\text{CO}_2 \text{ validation}$. The laboratory offset applied was $42.7 \mu\text{atm}$ (without peaks). The laboratory calibration applied to the data was that derived from the linear regression of data from the initial laboratory evaluation: $p\text{CO}_2 \text{ SIPCO2 corrected} = (p\text{CO}_2 \text{ SIPCO2 measured} - 22)/1.03$.

	SIPCO2 uncorrected data	SIPCO2 laboratory offset applied	SIPCO2 laboratory calibration applied
Off _{mean} (μatm)	29 (± 6)	-13.8 (± 6)	-5 (± 5)
Slope	1.08	1.04	1.04
Intercept	-1.9	-23.5	-23
R ²	0.97	0.97	0.97

at the CML site. First, we used a simple offset removal, subtracting $42.7 \mu\text{atm}$ from the SIPCO2 sensor readings at the CML (Table 4). This offset correction lowered the mean CML offset to $-13.8 \pm 6 \mu\text{atm}$, which is closer to an ideal offset of $0 \mu\text{atm}$. Second, we used the results of the linear regression of the laboratory evaluation and applied the same slope and intercept to the CML SIPCO2 sensor data (Table 4), according to Eq. 3:

$$p\text{CO}_2 \text{ corrected} = \frac{(p\text{CO}_2 \text{ SIPCO2} - 22)}{1.03} \quad (3)$$

This simple application of the slope and intercept from linear regression of the laboratory evaluation data produced the best SIPCO2 results from the CML (Fig. 4), bringing the mean offset to $-5 \pm 5 \mu\text{atm}$.

Discussion

For many pCO₂ measurement applications in estuaries, lakes, or rivers, a SIPCO2 accuracy and precision of $29 \pm 6 \mu\text{atm}$, as determined in the CML evaluation, or $42.7 \pm 14.2 \mu\text{atm}$ in the case of the laboratory evaluation, may be acceptable for examinations of air-water CO₂ fluxes or aquatic ecosystem metabolism; however, it is important to acknowledge that pCO₂ levels and ranges of variation in freshwater and estuary systems can be several times to an order of magnitude larger than the range tested in the CML evaluation. For instance, Abril et al. (2006) showed precision of directly measured river pCO₂ of $\pm 15\%$ over a 0–13,000 ppm range. Borges et al. (2015) documented a method precision of $\pm 2\%$ for IRGA-based direct measurements of African rivers ranging in pCO₂ from 300 ppm to 16,942 ppm. Hanson et al. (2003) determined net ecosystem productivity via IRGA-based pCO₂

measurements in 25 lakes, with an estimated pCO₂ precision of $\pm 2\%$ over a 0–20,000 ppm range. Raymond et al. (1997) showed a precision of $\pm 8\%$ over a range of about 500–2000 ppm in the tidal Hudson River estuary. Although the SIPCO₂ level of accuracy and precision may be unacceptable for “climate” quality oceanic studies, the precision of results from the CML evaluation ($\pm 6 \mu\text{atm}$) exceeds the “weather” target uncertainty of $\pm 10 \mu\text{atm}$ ($\pm 2.5\%$ at 400 μatm pCO₂), but falls short of the “climate” target uncertainty of $\pm 2 \mu\text{atm}$ ($\pm 0.5\%$ at 400 μatm pCO₂, Newton et al. 2015). Furthermore, we have found that a pre- (and perhaps post-) deployment laboratory calibration may offer a way to improve the data quality from the SIPCO₂ sensor.

We anticipate that a longer laboratory evaluation period with larger and more gradual pCO₂ changes than shown here will produce a more robust regression calibration. Thus, while a simple offset subtraction improved the SIPCO₂ sensor readings at the CML site, the generation of a linear regression relationship against readings from a validation system appears to be a reasonable way to correct readings from this SIPCO₂ sensor. It is worth noting that the CML evaluation spanned pCO₂ levels of a fairly narrow range (340–500 μatm). Freshwater field evaluations and verification of the linearity of the SIPCO₂ sensor response are presently underway. Preliminary results indicate that the SIPCO₂ sensor works well in the elevated pCO₂ environment found in local rivers, where a wider range of pCO₂ is expected. This is consistent with the K-30 measurement specifications of Bastviken et al. (2015). The results of Bastviken et al. (2015) indicate that corrections for detector temperature and relative humidity offer additional ways to improve the quality of CO₂ readings from the K-30 NDIR detector, but we expect humidity impacts to be low for the SIPCO₂ sensor due to typical operation at $\sim 100\%$ humidity in continuous mode. While Bastviken et al. (2015) noted problems with condensation, no condensation was observed in the SIPCO₂ housing during these tests.

Potential sources of error and uncertainty

The consistent presence of an offset between SIPCO₂ readings and those of the validation pCO₂ system (+42.7 μatm in the laboratory evaluation, Table 2, and +29 μatm in the CML test, Table 4) suggest that there are additional sources of error in the SIPCO₂ system. If quantified, knowledge of these errors could be used to improve the SIPCO₂ data quality. Due to its simple design, the SIPCO₂ system has two components which can create potential sources of error: the air pump and the K-30 NDIR detector. Faults in the air pump would result in incomplete equilibration or slower response times. While the response of the SIPCO₂ lags that of the validation system (Fig. 2), an increase in this lag on the order of minutes may have produced a smaller offset, as shown by the difference between the Off_{mean} values with and without peaks shown in Table 2. The air pump used in

both the Laboratory and CML evaluations was not observed to have failed or exhibited reduced flow rates.

The K-30 NDIR detector is a more likely source of error in the SIPCO₂ design. We performed three tests on a new K-30 NDIR sensor: linearity of response, temperature response, and humidity response. This K-30 module was not the same as used in the laboratory and CML evaluations, but was the same model and was prepared and calibrated with UHP N₂ and 400 ppm CO₂ in the same manner as described above. Similar to Bastviken et al. (2015), we compared the response of the K-30 NDIR to that of a Li-cor LI-840 NDIR, the same highly linear model used in the laboratory and CML evaluations of the SIPCO₂ sensor. This response test was performed by mixing streams of ultrapure nitrogen and pure CO₂ with a valve to produce mixtures from 0 ppm to near-10,000 ppm CO₂, then flowing the mixed gas stream sequentially through the LI-840 and K-30 NDIR detectors, which were connected by short lengths of tubing. Results of this linearity test show that the response of the K-30 NDIR is linear ($r^2 = 0.9998$, Fig. 5). However, the slope of the regression line between K-30 and LI-840 NDIR detectors indicates that the K-30 NDIR underestimates xCO₂ at higher concentrations (slope = 0.93). This is a somewhat different finding than that presented by Bastviken et al. (2015), who showed a linear response from the K-30 NDIR but used a comparison NDIR (Los Gatos Research DLT100) that exhibits quenching effects and nonlinear behavior at higher CO₂ concentrations. At 350 μatm pCO₂, the approximate minimum reading during the CML evaluation, this translates to an offset of $-26 \mu\text{atm}$ (calculated according to Eq. 1), while at 500 μatm pCO₂ the calculated offset is $-36 \mu\text{atm}$. The fact that this response linearity offset is negative (i.e., LI-840 readings are higher than K-30 readings), while the overall Off_{mean} results from the laboratory and CML evaluations were positive indicates that another factor was more than offsetting the differences in response linearity between the two NDIR detectors. This finding indicates that calibration of a K-30 NDIR with mixtures of gases may not be sufficient to quantify the overall performance of a SIPCO₂ sensor, and comparison to a validation system such as the one discussed here is advisable for highly accurate results.

Other studies examining the K-30 NDIR have demonstrated that the detector is sensitive to temperature and humidity changes (Yasuda et al. 2012; Bastviken et al. 2015). To examine the temperature response of the K-30 NDIR, we first calibrated the sensor at 20°C using ultra-high purity (UHP) nitrogen and 400 ppm CO₂ in nitrogen. We then flowed either UHP nitrogen or a 515 ppm mixture of CO₂ in nitrogen through the K-30 NDIR while altering the temperature of the K-30 detector. The temperature was altered by placing the K-30 NDIR on a chilled laboratory hot plate, which has previously been kept in a freezer. Contact between the K-30 NDIR and the chilled hot plate lowered the temperature of the K-30 to 5–10°C; the hot plate was then slowly warmed to

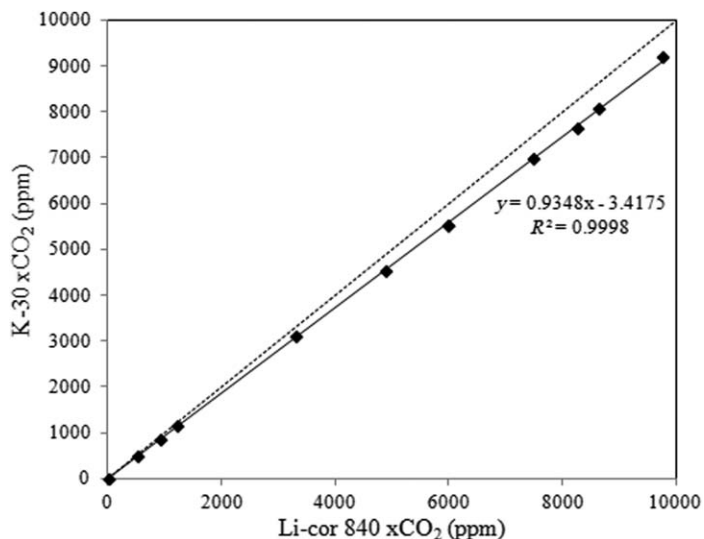


Fig. 5. Comparison of xCO₂ response between a Li-cor LI840 NDIR and a K-30 NDIR of the same model used in the SIPCO2 sensor. Standard deviation error bars in both the x- and y-directions are too small to appear in the plot. The dashed line indicates a 1 : 1 response, while the solid line shows the regression line between LI840 and K-30 detectors.

30–50°C, and the xCO₂ response of the K-30 NDIR was recorded (Fig. 6). The change in xCO₂ reading from temperature change was nearly identical between UHP nitrogen and 515 ppm CO₂. xCO₂ readings from the K-30 NDIR increased at temperatures greater than the calibration temperature (20°C), and fell below the known gas concentration at temperatures less than 20°C.

The K-30 NDIR has an onboard temperature sensor, and provides a detector temperature reading as part of its typical data stream. This offers the potential opportunity to temperature-correct the xCO₂ readings from a K-30 NDIR, using the logarithmic regression equation presented in Fig. 6. While Bastviken et al. (2015) noted a lack of temperature effect on K-30 readings, the equation shown in Fig. 6 is quite similar to the linear correction presented in Yasuda et al. (2012), a study in which several identical K-30 sensors were evaluated simultaneously.

To test the effect of varying humidity on K-30 NDIR readings, we employed a Li-cor 610 portable dew point generator. UHP nitrogen and the same 514 ppm CO₂ mixture employed above were flowed through the LI-610 instrument set to a dew point of 0°C, then subsequently to the K-30 NDIR sensor. The LI-610 was then adjusted to gradually increase the dew point of the UHP nitrogen or 514 ppm CO₂ mixture, while readings from the K-30 sensor were recorded. This test showed increasing xCO₂ readings from the K-30 sensor for UHP nitrogen, while the 514 ppm CO₂ mixture showed an initial slight xCO₂ decrease, followed by an xCO₂ increase above a dew point of about 10°C (Fig. 7). The temperature of the K-30 NDIR did not change appreciably over

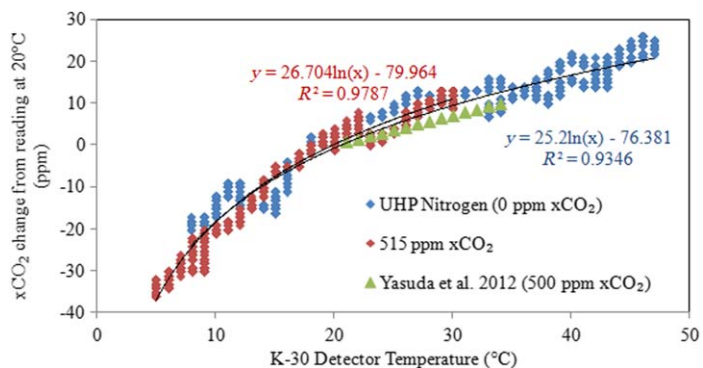


Fig. 6. Change in K-30 NDIR readings of a constant sample gas stream with varying temperature. Changes in UHP N₂ gas (0 ppm xCO₂) are shown in blue, while those of a 515 ppm CO₂/N₂ mixture are shown in red. The K-30 NDIR was calibrated at 20°C. Results from Yasuda et al. (2012) are shown in green, also for a K-30 NDIR calibrated at 20°C. The black line shows a logarithmic regression of the combined UHP N₂ and 515 ppm CO₂/N₂ data.

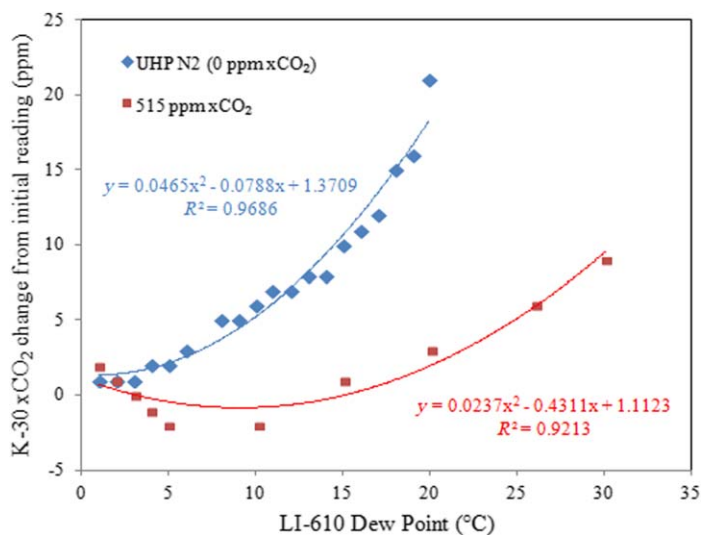


Fig. 7. Response of a K-30 NDIR detector to humidity change produced by a Li-cor LI-610 dew point generator. Blue points show changes in xCO₂ readings for UHP N₂ gas (0 ppm xCO₂), while red points show xCO₂ changes for a 515 ppm CO₂/N₂ mixture. Second-order polynomial regression lines and equations are shown in blue and red for the UHP N₂ gas and 515 ppm CO₂/N₂ mixture, respectively.

either dew point test, indicating that changes in xCO₂ readings were due solely to humidity. The manufacturer of the K-30 NDIR also offers models with onboard relative humidity measurements (i.e., K-33 ELG), the use of which offer the potential to correct for both temperature and humidity effects. While these models are somewhat more expensive, the capability to perform both temperature and humidity corrections would likely be valuable, particularly if the SIPCO2 sensor is to be deployed outside under a wide range of environmental conditions.

In light of these error sources, we can speculate whether some combination of the systematic K-30 underestimate of xCO₂ (Fig. 5), temperature effects (Fig. 6), and humidity effects (Fig. 7) can explain much of the Off_{mean} determined in both the laboratory and CML evaluations (Tables 2, 4). Unfortunately, the temperature detector in the SIPCO2 K-30 used in both laboratory and CML evaluations malfunctioned, making post-correction of the data impossible. However, we can estimate what the potential effects would be at some likely temperatures, with a few assumptions. First, we assume that the air inside the SIPCO2 was at 100% humidity, as it was constantly recirculated beneath the water's surface by the air pump while remaining isolated from surrounding ambient air. Second, both the laboratory and CML evaluations were performed inside temperature-controlled buildings (although the CML building was generally colder), so we assume the air temperature inside the SIPCO2 was between 15°C and 25°C. Given these assumptions, the contribution of humidity to the observed Off_{mean} should have been small (0 μatm at 15°C to 5 μatm at 25°C). Similarly the temperature contribution to the observed Off_{mean} should also have been small (−8 μatm at 15°C to 5 μatm at 25°C, given a K-30 calibration temperature of 20°C). As previously discussed, at the minimum CML evaluation pCO₂ of 350 μatm the linearity offset is −26 μatm (calculated according to Eq. 1), while at 500 μatm pCO₂ the offset is −36 μatm. Combining linearity and humidity errors yields a net error offset ranging from −44 μatm (at 15°C and 500 μatm) to −16 μatm (at 25°C and 350 μatm). Under the above assumptions, the detector temperature would have to be 183°C to account for the linearity error, humidity error, and observed Off_{mean} at 350 μatm, while the temperature would have to be 284°C to account for the same errors at 500 μatm. These results indicate that factors other than linearity, temperature, and humidity are producing the observed Off_{mean}; the results also underscore the importance of a careful evaluation of a SIPCO2 system against a validation system before attempting to make accurate measurements.

Comparison to other pCO₂ sensors

The pCO₂ readings from the SIPCO2 sensor before post-deployment adjustment underperformed some commercial sensors. For example, Degrandpre et al. (1995) demonstrate a SAMI-CO₂ accuracy/precision of ±2 μatm and ±1 μatm, respectively, although an intercomparison with a NDIR-based equilibrator system resulted in an overall offset between the two systems of 3.7 μatm. In comparison to a NDIR-based equilibrator system, Fietzek et al. (2013) found a Contros Hydro-C offset and precision of −0.6 μatm and ±3 μatm, respectively. However, after careful characterization against a validation system the SIPCO2 sensor performance is comparable to that of some of the systems presented in Table 1. The sensors discussed in the ACT evaluations: Contros Hydro-C (Fietzek et al. 2013), Pro-Oceanus CO₂-Pro,

Sunburst SAMI-CO₂ (Degrandpre et al. 1995), and PMEL/Batelle MAPCO2 (Friederich et al. 1995) were tightly restricted, while we had continuous access to the SIPCO2 system, so this work does not present an evaluation conducted under directly comparable conditions to many of the sensors listed in Table 1.

The presented SIPCO2 sensor configuration (\$200USD to \$500USD for sensor components, see Supporting Information Material) offers a significant price advantage in comparison to the commercial systems described in Table 1. One disadvantage of the SIPCO2 sensor is an assembly time of several hours combined with a further time requirement to evaluate sensor performance, as opposed to an off-the-shelf system. However, we also believe the SIPCO2 configuration, using a small air pump to recirculate air within a small housing, offers a reduced risk of biofouling or sensor damage when compared to sensors that employ a membrane to exchange CO₂. Another advantage of the SIPCO2 design is the modular nature of the sensor, which allows for the easy replacement of components that may fail over time.

One disadvantage of the SIPCO2 sensor as described here is its response time, which is somewhat slow in comparison to other commercial sensors (on the order of 10 min in the laboratory evaluation). The use of a more powerful air pump to provide a higher flow rate should mitigate this issue, although we have not tested this possibility. The results in this work suggest that the SIPCO2 sensor could support many fixed-location scientific and monitoring applications, as opposed to use as a portable handheld pCO₂ sensor such as described in Johnson et al. (2010). Interestingly, Bastviken et al. (2015) show a negligible difference in pCO₂ readings between pumped and unpumped systems, although their measurement timescale was on the order of several hours. Further investigation into response time and mixing rates is warranted depending on specific application needs.

A key issue in the use of the SIPCO2 sensor relates to calibration. The host software for the K-30 CO₂ detector used in the SIPCO2 sensor at present only allows calibrations with gases of zero and 400 ppm CO₂. This is a relatively narrow calibration range for many applications, as pCO₂ can range an order of magnitude or higher in aquatic systems (Raymond et al. 1997; Hunt et al. 2013; Bastviken et al. 2015). While Bastviken et al. (2015) and results in this work show good K-30 sensor linearity, this limited K-30 calibration range underscores the need to characterize the SIPCO2 system against a validation system in a variety of environments.

Because the SIPCO2 sensor housing is not completely sealed from the environment, the present design is not submersible, and thus is only useful for surface measurements or integrated measurements of pCO₂ down to the depth of the pump's air tubing. Furthermore, this version has no ability to log data internally or operate autonomously on battery power. However, the cost, ease of assembly, and quality of

data suggest that the SIPCO2 provides a potentially valuable instrument for observation nodes in freshwater, estuary, and coastal environments. The addition of low-cost components could readily address data-logging and power issues. The SIPCO2 sensor represents a basic platform for aquatic pCO₂ measurements, to which users could add functionality to suit individual needs and requirements.

Comments and recommendations

This paper presents a new design for a relatively inexpensive sensor to measure aquatic pCO₂ from readily obtainable components, evaluates the performance of the sensor in two settings, and provides a simple method to correct the sensor data against a known validation system. Uncorrected SIPCO2 sensor performance is below that of comparable commercial sensors ($30 \pm 6 \mu\text{atm}$ difference from validation system), but post-deployment correction of the SIPCO2 measurements produces data of comparable quality ($-5 \pm 5 \mu\text{atm}$ difference from validation system).

While many researchers may prefer out-of-the-box sensors for their application, we expect that the open design and cost savings of the SIPCO2 sensor will make it appealing to many users. The supplementary material provides a list of SIPCO2 sensor components, suppliers for many of the components, as well as step-by-step instructions for assembly. We strongly recommend calibration of the SIPCO2 sensor against a known validation system in order to produce the highest quality data possible. One potentially powerful use of this sensor is within a network of observation nodes, where numerous sensors could be deployed spatially around a validation system. This could dramatically increase the extent of in situ pCO₂ coverage while assuring data quality using a single expensive validation system.

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Conflict of Interest

None declared.

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